

(1) (a) $L_{\mathbb{H}}(1|1) = \frac{1}{2}$ $L(2|1) = \frac{1}{3}$ (1)

$L_{\mathbb{H}}(1|2) = \frac{1}{3}$ $L(2|2) = \frac{4}{5}$

$L_{\mathbb{H}}(1|3) = \frac{1}{2}$ $L(2|3) = \frac{1}{6}$

and any positive multiple of these gives all possible profile likelihoods.

(b) $C_{1/2}(1) = \{ \tau : L_{\mathbb{H}}(\tau|1) \geq \frac{1}{2} \sup_{\tau} L(\tau|1) \} = \{ \tau : L_{\mathbb{H}}(\tau|1) \geq 1/4 \} = \{1, 2, 3\}$

$C_{1/2}(2) = \{ \tau : L_{\mathbb{H}}(\tau|2) \geq \frac{2}{5} \} = \{2, 3\}$

$C_{1/2}(3) = \{ \tau : L_{\mathbb{H}}(\tau|3) \geq \frac{1}{4} \} = \{1, 3\}$

(2) (a) $L(\mu, \sigma^2 | x) = (\sigma^2)^{-\frac{n+1}{2}} \exp\left\{ -\frac{n}{2\sigma^2} (\bar{x} - \mu)^2 - \frac{(n+1)s_x^2}{2\sigma^2} \right\}$

and $L(\tau, \sigma^2 | x) = (\sigma^2)^{-\frac{n+1}{2}} \exp\left\{ -\frac{n}{2\sigma^2} (\bar{x} - \tau)^2 - \frac{(n+1)s_x^2}{2\sigma^2} \right\}$

and note that $\tau \in \mathbb{R} \setminus \{0\}$, $\sigma^2 > 0$ are unconstrained so $L_{\mathbb{H}}(\tau | x) = \sup_{\sigma^2 > 0} L(\tau, \sigma^2 | x)$ and we need to maximize $L(\tau, \sigma^2 | x)$ as a function of σ^2 .

(b) Taking the log of the likelihood we obtain

$l(\tau, \sigma^2 | x) = -\frac{n+1}{2} \log \sigma^2 - \frac{n}{2\sigma^2} (\bar{x} - \tau)^2 - \frac{(n+1)s_x^2}{2\sigma^2}$ and

$$\frac{\partial \ell(\tau, \sigma^2 | x)}{\partial \sigma^2} = -\frac{n}{2\sigma^2} + \frac{n}{2\sigma^4} (\bar{x} - \frac{\sigma^2}{\tau})^2 + \frac{n}{2\sigma^2} 2(\bar{x} - \frac{\sigma^2}{\tau}) \frac{1}{\tau} (\sigma^2)^{\frac{1}{2}} + \frac{(n+1)\sigma^2}{2\sigma^4 \tau}$$

Then setting this equal to 0 and multiplying through by $-2\sigma^4/n$ we get

$$\sigma^3 + \sigma(\bar{x}^2 - 2\bar{x}\sigma/\tau + \sigma^2/\tau^2) - \sigma^3(\bar{x} - \sigma/\tau) \frac{1}{\tau} - \frac{(n+1)\sigma^2}{2\tau} = 0$$

$$= \sigma^3 - 2\bar{x}\sigma + \frac{\bar{x}^2}{\tau}\sigma^2 - \frac{\sigma^3}{\tau} - \frac{\bar{x}\sigma^2}{\tau} + \frac{\sigma^3}{\tau} - \frac{(n+1)\sigma^2}{2\tau} = 0$$

$$= \sigma^3 + \frac{\bar{x}^2}{\tau}\sigma^2 - (\bar{x} + \frac{n+1}{2\tau}\sigma) \sigma = 0$$

$$\text{or } \sigma^2 + \frac{\bar{x}^2}{\tau}\sigma - (\bar{x} + \frac{n+1}{2\tau}\sigma) = 0$$

$$\text{so } \sigma = \frac{-\frac{\bar{x}^2}{\tau} + \sqrt{\frac{\bar{x}^4}{\tau^2} + 4(\bar{x} + \frac{n+1}{2\tau}\sigma)}}{2}$$

$$= \frac{1}{2} \left(-\frac{\bar{x}^2}{\tau} + \sqrt{\frac{\bar{x}^4}{\tau^2} + 4\bar{x} + \frac{2(n+1)\bar{x}\sigma}{\tau}} \right) \text{ since } \frac{\bar{x}^4}{\tau^2} + 4\bar{x} + \frac{2(n+1)\bar{x}\sigma}{\tau} = \left(\frac{\bar{x}^2}{\tau} + 2\bar{x} + \frac{(n+1)\sigma}{\tau} \right)^2$$

and since σ is not negative the root must be

$$\sigma(\tau) = \frac{1}{2} \left(-\frac{\bar{x}^2}{\tau} + \sqrt{\frac{\bar{x}^4}{\tau^2} + 4\bar{x} + \frac{2(n+1)\bar{x}\sigma}{\tau}} \right)$$

Then the profile likelihood f_n is

$$L_{\tau}(\tau | x) = (\sigma^2(\tau))^{\frac{n+1}{2}} \exp\left\{ -\frac{n}{2\sigma^2(\tau)} \left(\bar{x} - \frac{\sigma^2(\tau)}{\tau} \right)^2 - \frac{(n+1)\sigma^2(\tau)}{2\sigma^2(\tau)} \right\}$$